

Model Tuning

Session Plan



- 1. Introduction
- 2. Discussion Questions on the concepts
- 3. Hands-on Case study
- Extended Discussions and QnA
- 5. Summary

Discussion Questions

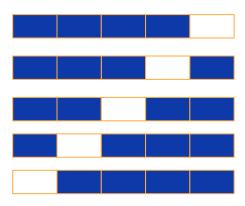


- 1. What are the steps of cross-validation?
- How to handle imbalanced data?
- 3. What is Data leakage?
- 4. How to deal with the situation where the model shows underfitting?
- 5. How to deal with the situation where the model shows overfitting?
- 6. Missing value imputation using KNN imputer



What are steps involved in cross-validation?

- Cross-validation is a technique used for evaluating models
- K fold cross-validation will divide data into k-folds
- Train model on k-1 folds and test its performance on the last fold
- K fold cross-validation will generate k models and k performance scores
- Instead of getting only 1 score, here we'll get k scores, which will give a better picture of the variance in model performance



How to handle imbalanced data?



- Datasets used in banking, health and market analytics usually have imbalance i.e. one class is in majority and one is in minority (less than 5%)
- During training on such datasets, the model gives more weightage to the majority class and gets biased
- To avoid such situations, we can use oversampling or undersampling techniques on data
- Oversampling will create artificial data points for the minority class
- Undersampling will remove data points from the majority class
- We can't afford to lose data points in case of small data size, so oversampling is preferred in such cases

Data leakage



- Data leakage is the situation where the model, while it is being created, is influenced by test data
- Due to data leakage, model performance on test data is not trustworthy as the sanctity of test data is compromised
- Data leakage can happen in multiple ways.
 - Standardizing data before splitting into training and testing data. For e.g. using z-score
 - Imputing missing values for the entire data before splitting into training and testing data
 - Hyper parameter tuning to improve performance on test data
- Best way to avoid data leakage is to keep a portion of the sample data away before doing any processing

What is underfitting?



- We say a model is underfitting when it is not performing well on the train set
- This situation arises when a model is not able to learn from the train set

Reasons for underfitting

Small data size with a large number of features

Less model complexity

Irrelevant features

Imbalanced data

Dealing with underfitting

Increase model complexity, i.e. if you were using only a linear combination of features then try using a non-linear combination

In case of imbalanced data, use oversampling or undersampling

In the case of small-sized datasets with a large number of features, use features that seem important as per the need

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What is overfitting?



- We say a model is overfitting when it performs good on train data but not good enough on test/unseen data
- This situation arises when the model starts learning the noise and inaccurate data entries

What could be the reason for this?

- High model complexity
- Small dataset
- Noisy data

Train accuracy = 98.01!!



Test accuracy = 55.87



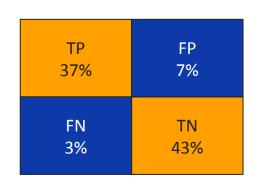
How to detect overfitting?

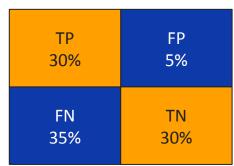


- Check model performance on train set and test set if there is a huge difference in both, then we can say that model is overfit
- But sometimes we might get a biased train-test split i.e., train data has different distribution as compared to the test set
- So to confirm if we truly have overfitting or not, one must check model performance via cross validation

Dealing with overfitting

- Regularization
- Train with more data
- Remove irrelevant features
- Decrease model complexity





Confusion matrix on train

Confusion matrix on test

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K- nearest neighbours



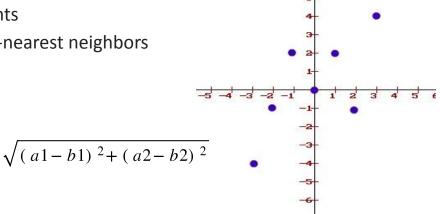
Before understanding how KNN Imputer works, let's understand what are K-nearest neighbors (KNN)

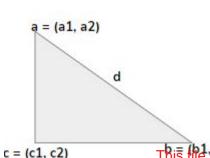
Looking at the graph what are the 4-nearest neighbors of (0, 0)?

- 1. Find the euclidean distance of (0, 0) from all other points
- 2. 4 Points with the least Euclidean distance will be the 4-nearest neighbors

How to calculate Euclidean distance?

The Euclidean distance between **a** and **b** is **d** and d is:





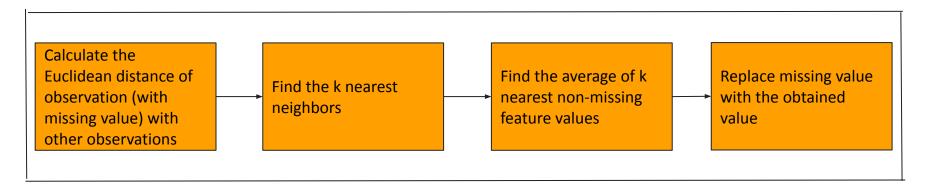
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KNN Imputer



How does KNN Imputer works?

KNN imputer replaces missing values using the average of k nearest non-missing feature values (k needs to be decided by us)

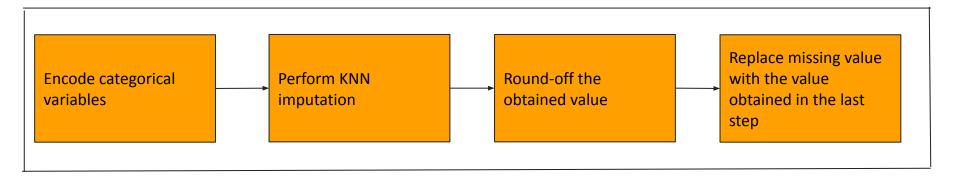




KNN Imputer for categorical data

Since missing values are getting replaced with average values - How to do imputation in the case of a categorical variable?

Missing values in categorical data should be replaced with the nearest integer obtained via KNN imputer





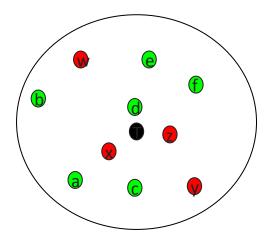
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KNN Imputer for categorical data - example

- Looking at the image, assume red and green be 2 different categories of a variable
- And the black ball 'T', has this variable missing and we want to impute that

The process to calculate missing category of 'T'

- 1. Encode green as 0 and red as 1
- 2. Consider k = 3, find out 3 nearest neighbors
- 3. We can see that d, x and z are 3 nearest neighbors of T
- 4. d has the category encoded as 0 and x, z as 1
- 5. So the average value is (0+1+1)/2 = 0.66
- 6. Rounding-off 0.66 we get 1, so the category assigned to 'T' is 1
- 7. Reverse encode the categories, so 'T' will be assigned the red category



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Happy Learning!

